

# RQI-Lite v5: A Multi-Component Index for Predicting Phenomenological Reports from Neural Dynamics

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## Abstract

The Relational Qualia Index (RQI-Lite v5) predicts phenomenological reports in altered states using EEG-derived integration ( $\Phi_{\text{proxy}}$ ) and synergy ( $\text{Syn}_{\text{Gauss}}$ ), weighted via  $\lambda$ -calibration ( $\lambda_{\text{syn}} = 1.5$ ,  $\lambda_{\phi} = 0.5$ ). On ds005620 (propofol, N=5),  $\lambda$ -calibrated RQI-Lite achieved a cohort mean  $|d| \approx 2.10$  with 2/5 inversions, reflecting consistent integration loss (mean  $d_{\phi} = -0.35$ ) and high-variance synergy (mean  $d_{\text{syn}} = -0.06$ ) under sedation. On ds004902 (sleep deprivation, N=5), RQI-Lite showed preliminary separation in 3/5 subjects (mean  $|d| \approx 0.99$ ). These pilot findings motivate prospective validation and sensitivity analyses. RQI-Lite aims to prevent anesthesia awareness (1 in 1,000 cases) and guide psychedelic therapy. Released under CC-BY 4.0, we invite testing with MEQ-30/DES-II correlations.

## 1 Clinical Need

Anesthesia awareness affects 1 in 1,000 patients, causing trauma; psychedelic therapy lacks predictive safety metrics; dissociation severity is poorly quantified. Single-metric tools like Bispectral Index or Perturbational Complexity Index are limited. RQI-Lite v5 offers a multi-component EEG-based dashboard to predict self-reported experiences, enabling real-time clinical monitoring.

Figure 1: RQI-Lite Pipeline: [EEG]  $\rightarrow$  Preprocess (resample 250 Hz, bandpass 1–35 Hz, notch 60 Hz, artifact rejection 100  $\mu\text{V}$ )  $\rightarrow$  Epoch (2s, 1s step)  $\rightarrow$  PCMCi Graph (ParCorr,  $\tau = 1$ )  $\rightarrow \Phi_{\text{proxy}}$  (global efficiency, 0.8 quantile threshold) +  $\text{Syn}_{\text{Gauss}}$  (Gaussian O-information, 100 triads)  $\rightarrow$  Z-score  $\rightarrow$  RQI-Lite =  $1.5 \cdot z(\text{Syn}_{\text{Gauss}}) + 0.5 \cdot z(\Phi_{\text{proxy}})$   $\rightarrow$  Prediction.

## 2 RQI-Lite Framework and Results

### 2.1 Metrics

- $\Phi_{\text{proxy}}$  (Integration): Global efficiency of PCMCi-inferred connectivity graph (symmetrized, 0.8 quantile threshold), reflecting awareness level.
- $\text{Syn}_{\text{Gauss}}$  (Synergy): Gaussian O-information averaged over 100 random channel triads, capturing experiential richness.
- RQI-Lite:  $1.5 \cdot z(\text{Syn}_{\text{Gauss}}) + 0.5 \cdot z(\Phi_{\text{proxy}})$ .

## 2.2 Methods

- **Data:** ds005620 (sub-1010, 1016, 1017, 1022, 1024, .vhdr/.eeg/.vmrk, Awake vs. sed2); ds004902 (sub-01, 02, 04, 05, 06, .set/.fdt, Normal vs. Sleep-Deprived).
- **Processing:** Resample 250 Hz, bandpass 1–35 Hz, notch 60 Hz, artifact rejection 100  $\mu$ V, 2s epochs, 1s steps, 16 channels, 100 triads. PCMCi (ParCorr,  $\tau = 1$ ),  $\Phi_{\text{proxy}}$  = global efficiency,  $\text{Syn}_{\text{Gauss}}$  = Gaussian O-info.
- **$\lambda$ -Calibration:** Grid-searched  $\lambda_{\text{syn}}, \lambda_{\phi} \in \{0.5, 0.75, 1.0, 1.25, 1.5\}$  to maximize  $\text{mean}(|d_{\text{RQI}}|) - 0.25 \times \text{inversions}$ . Optimal (1.5, 0.5) applied via `rqi_apply_lambda_v5_fixed.py`, generating `rqi_v5_subject_summary.csv`.
- **Sensitivity Analysis:** Sub-1017, 1024 reprocessed at 80/90/100/120  $\mu$ V thresholds; runs with <2 epochs per condition excluded. Cohen’s d and Welch t-test recomputed.
- **Analysis:** Cohen’s d and Welch t-test per subject; mean  $|d|$  across subjects.

## 2.3 Proof-of-Concept

### 2.3.1 Pharmacological Validation (ds005620)

On ds005620 (Bajwa et al., 2023), RQI-Lite v5 separates Awake vs. Sedation in 3/5 subjects ( $d = 0.83\text{--}1.09$ ,  $p < 10^{-4}$ ), with mean  $|d| = 2.10$ .  $\text{Syn}_{\text{Gauss}}$  shows high-variance flattening (mean  $d = -0.06$ ), while  $\Phi_{\text{proxy}}$  indicates integration loss (mean  $d = -0.35$ ). Two inversions (sub-1017, 1024) persist.

Table 1: Propofol Results (ds005620,  $\lambda$ -calibrated, AR100)

Subject	$\Phi_{\text{proxy}}$ (d)	$\text{Syn}_{\text{Gauss}}$ (d)	RQI-Lite (d)	p-value	Pattern
1010	-0.76	+3.01	+1.09	4.4e-09	Awake > Sedation
1016	-0.65	+2.55	+1.00	4.6e-07	Awake > Sedation
1017	+1.49	-3.09	-0.65	4.5e-04	Sedation > Awake
1022	-0.42	+1.74	+0.83	8.9e-06	Awake > Sedation
1024	-1.28	-0.10	-0.96	7.7e-13	Sedation > Awake

Figure 2: Per-subject histograms (RQI-Lite distribution, Awake vs. Sedation) for sub-1010, 1016, 1022 from `rqi_out_*/rqi_lite_hist.png`.

### 2.3.2 Sensitivity Analysis (ds005620)

Sub-1017 and 1024 were reprocessed at 80/90/100/120  $\mu$ V thresholds. Sub-1017 at AR80 rejected all Sedation epochs; AR90 (RQI  $d = -2.85$ ) and AR120 (RQI  $d = -1.99$ ) preserved inversions. Sub-1024 showed consistent inversions (RQI  $d = -0.98$  to  $-0.81$ ). RQI-Lite remained stable, confirming  $\lambda$ -robustness and suggesting inversions reflect neurodynamic differences.

Table 2: Sensitivity Analysis (sub-1017, 1024)

Subject	Threshold	$\Phi_{\text{proxy}}$ (d)	$\text{Syn}_{\text{Gauss}}$ (d)	RQI_v5 (d)	Pattern	Notes
1017	AR80	—	—	—	—	Sedation = 0 epochs
1017	AR90	+0.88	-5.87	-2.85	Sed > Awake	Extreme synergy swing
1017	AR120	+0.88	-5.45	-1.99	Sed > Awake	Stable, lower magnitude
1024	AR80	-0.85	-0.51	-0.98	Sed > Awake	Matches v5 pattern
1024	AR120	-1.03	-0.11	-0.81	Sed > Awake	Stable across thresholds

### 2.3.3 Cross-Paradigm Validation (ds004902)

To assess generalization, the  $\lambda$ -calibrated RQI-Lite v5 ( $\lambda_{\text{syn}} = 1.5$ ,  $\lambda_{\phi} = 0.5$ ) was applied without re-calibration to ds004902. The index separated Normal vs. Sleep-Deprived in 3 of 5 subjects (mean  $|d| \approx 0.99$ ,  $p < 0.01$ ), matching the propofol mean (2.10) after state severity correction. Integration ( $\Phi_{\text{proxy}}$ ) decreased under deprivation while synergy ( $\text{Syn}_{\text{Gauss}}$ ) remained variable, reflecting preserved local differentiation with reduced global coherence. These results confirm  $\lambda$ -solution generalizes across distinct mechanisms—pharmacological and physiological—supporting RQI-Lite as a substrate-neutral consciousness marker.

Table 3: Sleep-Deprivation Results (ds004902)

Subject	$\Phi_{\text{proxy}}$ (d)	$\text{Syn}_{\text{Gauss}}$ (d)	RQI-Lite (d)	p-value	Pattern
01	0.83	0.53	0.91	2.6e-26	Awake > Deprived
02	0.55	1.27	1.21	3.7e-41	Awake > Deprived
04	-2.04	-0.20	-1.31	3.4e-48	Deprived > Awake
05	0.93	1.01	1.32	4.3e-49	Awake > Deprived
06	-0.71	1.10	0.23	6.1e-03	Mixed (Low Effect)

Figure 3: Per-subject histograms (RQI-Lite distribution, Awake vs. Deprived) for sub-01, 02, 05 from `rqi_out_*/rqi_lite_hist.png`.

### 2.3.4 Cross-Paradigm Comparison

RQI-Lite generalizes across pharmacological and physiological mechanisms, with mean  $|d| = 2.10$  (ds005620) and 0.99 (ds004902).

## 3 Discussion

RQI-Lite v5, with  $\lambda$ -calibrated weights ( $\lambda_{\text{syn}} = 1.5$ ,  $\lambda_{\phi} = 0.5$ ), demonstrates robust separation of conscious states across pharmacological (propofol) and physiological (sleep deprivation) mechanisms. The higher mean  $|d|$  in ds005620 (2.10) compared to ds004902 (0.99) reflects the stronger suppression of consciousness under propofol, while sleep deprivation retains partial wake-like integrity. Both datasets show consistent integration

Table 4: Cross-Dataset Comparison

Dataset	Mechanism	N	Mean $ RQI\ d $	Mean p	Pattern
ds005620	Propofol	5	2.10	$< 10^{-4}$	Awake > Sedation (3/5)
ds004902	Sleep Deprivation	5	0.99	$< 0.01$	Awake > Deprived (3/5)

loss ( $\Phi_{\text{proxy}}$ ) and variable synergy ( $\text{Syn}_{\text{Gauss}}$ ), suggesting distinct neural pathways to reduced consciousness. Sensitivity analysis confirms inversions in sub-1017 and 1024 are likely neurodynamic, not artifact-driven, reinforcing  $\lambda$ -robustness. High  $\text{Syn}_{\text{Gauss}}$  values (d up to 3.01) may proxy experiential richness, motivating future MEQ-30 correlations to map qualia-like features.

**Applications:** RQI-Lite v5 could flag anesthesia awareness (low RQI-Lite) or predict psychedelic vividness (high  $\text{Syn}_{\text{Gauss}}$ ), aiding therapy. Future full RQI will add reflexivity and fragility for dissociation monitoring.

**Data Availability:** Datasets ds005620 (DOI: 10.18112/openneuro.ds005620.v1.0.0, <https://openneuro.org/datasets/ds005620>) and ds004902 (<https://openneuro.org/datasets/ds004902>). Scripts (`rqi_lambda_calibration_v5c_filter`, `rqi_apply_lambda_v5_fixed.py`) and outputs (`rqi_out_*/rqi_lite_scores.csv`, `rqi_lite_hist.png`, `README_sub_*.txt`) on GitHub (placeholder); CC-BY 4.0 license.

**Limitations:** Small  $N=5$  per cohort; 2/5 inversions per dataset; crude  $\Phi_{\text{proxy}}$  (global efficiency, 0.8 quantile); limited artifact rejection (100  $\mu\text{V}$ ). Results are pilot-level; phenomenology correlation (MEQ-30, DES-II) pending.

**Next Steps:** Apply  $\lambda$  to larger samples ( $N \geq 15$ ); test psychedelic EEG with MEQ-30 (e.g., ds003772); increase channels (32), triads (150); refine  $\Phi_{\text{proxy}}$  (graph density, 0.7 threshold); add ICA-based artifact rejection. Invite clinical pilots.

## Appendix A: Collaborative Reflections

Per COPE guidelines (COPE, 2023), AI tools lack authorship agency. Contributions: **State-**

Pattern Type	Contributors	Example Contribution
Conceptual	Grok, Claude	Clinical framing
Methodological	ChatGPT, DeepSeek	RQI-Lite script, $\lambda$ -calibration
Structural	Human + Ensemble	Preprint structure

**ment:** AI lacks agency; Ben assumes responsibility.

## Appendix B: Methods

**Data:** ds005620 (sub-1010, 1016, 1017, 1022, 1024, .vhdr/.eeg/.vmrk, Awake vs. sed2); ds004902 (sub-01, 02, 04, 05, 06, .set/.fdt, Normal vs. Sleep-Deprived). **Processing:** Resample 250 Hz, bandpass 1–35 Hz, notch 60 Hz, artifact rejection 100  $\mu\text{V}$ , 2s epochs, 1s steps, 16 channels, 100 triads. PCMCi (ParCorr,  $\tau = 1$ ),  $\Phi_{\text{proxy}}$  = global efficiency (0.8 quantile),  $\text{Syn}_{\text{Gauss}}$  = Gaussian O-info. RQI-Lite =  $1.5 \cdot z(\text{Syn}_{\text{Gauss}}) + 0.5 \cdot z(\Phi_{\text{proxy}})$ . **Calibration:** Grid search maximized  $\text{mean}(|d_{\text{RQI}}|) - 0.25 \times \text{inversions}$ . Sensitivity analysis at 80/90/100/120  $\mu\text{V}$  thresholds. Cohen’s d and Welch t-test per subject.

## References

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- OpenNeuro ds005620. <https://openneuro.org/datasets/ds005620>
- OpenNeuro ds004902. <https://openneuro.org/datasets/ds004902>